



COMP90042

Web search and text analysis

Workshop Week 3



Your tutor

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- Here, you can find my workshop slides:
- <https://github.com/winnchow/COMP90042-Workshops>

Postings list

Inverted Index - Recap

Document frequency

Document IDs

| term t | f_t | Postings list for t |
|----------|-------|-------------------------------------------------------------------------------------------------------|
| and | 6 | $\langle 1, 6, 7, 8, 9, 12 \rangle, \langle 1, 2, 1, 3, 1, 2 \rangle$ |
| big | 3 | $\langle 2, 5, 42 \rangle, \langle 1, 1, 1 \rangle$ |
| old | 1 | $\langle 32 \rangle, \langle 4 \rangle$ |
| in | 7 | $\langle 2, 3, 5, 6, 8, 14, 25 \rangle, \langle 1, 1, 4, 1, 5, 3, 1 \rangle$ |
| the | 52 | $\langle 1, 2, 3, 4, 5, 7, 8, 9, \dots \rangle, \langle 10, 21, 10, 42, 12, 14, 12, 4, \dots \rangle$ |
| night | 4 | $\langle 1, 12, 13, 14 \rangle, \langle 2, 2, 1, 3 \rangle$ |
| house | 5 | $\langle 6, 21, 32, 33, 43 \rangle, \langle 2, 3, 4, 2, 1 \rangle$ |
| sleep | 3 | $\langle 1, 51, 53 \rangle, \langle 1, 2, 3 \rangle$ |
| where | 4 | $\langle 1, 3, 4, 6 \rangle, \langle 1, 1, 2, 1 \rangle$ |

Term frequency

Compression

- How should we compress the document IDs?

| term t | f_t | Postings list for t |
|----------|-------|-----------------------------------------------------------------------|
| and | 6 | $\langle 1, 6, 7, 8, 9, 12 \rangle, \langle 1, 2, 1, 3, 1, 2 \rangle$ |

- Document IDs: $\langle 1, 6, 7, 8, 9, 12 \rangle$
- **Gaps**: $\langle 1, 5, 1, 1, 3 \rangle$, so mostly small numbers
- Variable Byte (Vbyte) Compression



Mostly small numbers

- For example, 1,1,1,1,1,1,1,1,1,1,1,1,1,1,2,3
- We may encode 1,2,3 using 2 bits each.
- How about we use $0 \Rightarrow 1$, $10 \Rightarrow 2$, $11 \Rightarrow 3$?

Variable Byte Compression

Examples

| Number | Encoding |
|--------|-------------------|
| 824 | 00111000 10000110 |
| 5 | 10000101 |



824 = 110 0111000

Storage Cost

| Number Range | Number of Bytes |
|-----------------|-----------------|
| 0 – 127 | 1 |
| 128 – 16383 | 2 |
| 16384 – 2097151 | 3 |

Q1 (c)



-
- Determine the values of integers X and Y that were encoded as the byte sequence [52,34,147,42,197] using the Variable Byte algorithm described in the lecture slides 9/10.

Q1 (c)

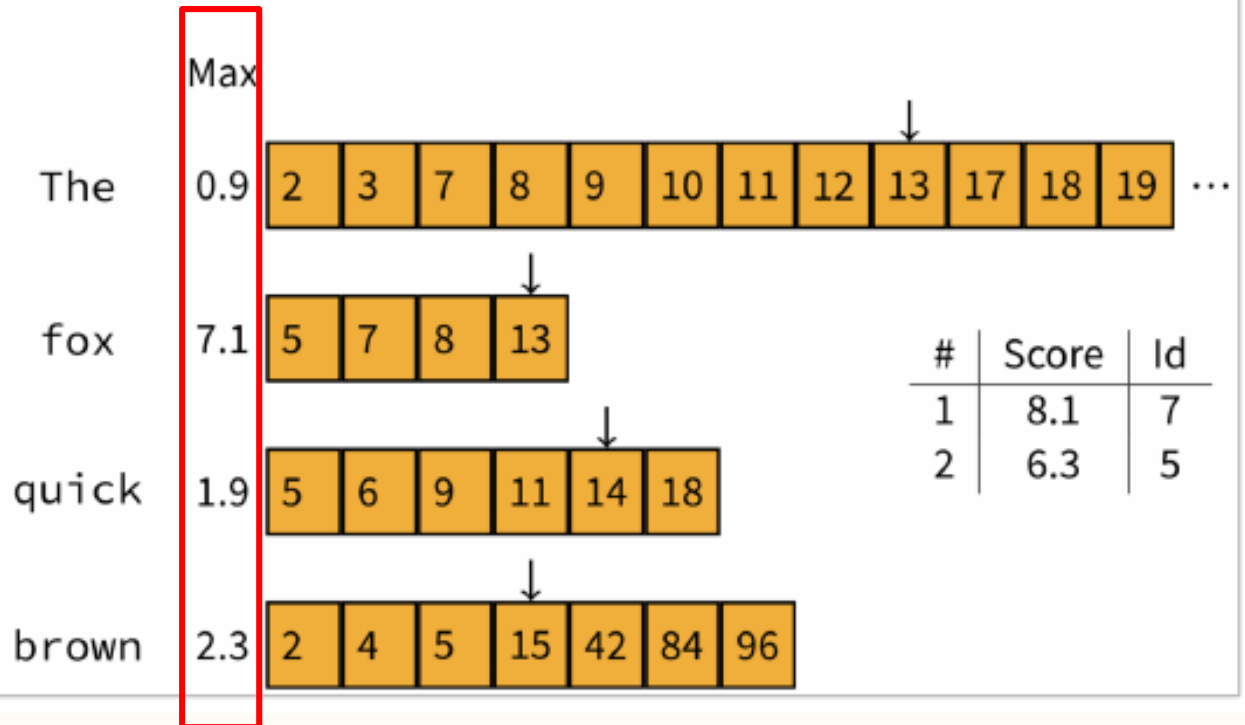
- 52 = 00110100
- 34 = 00100010
- 147 = 10010011
- => 0010011 0100010 0110100 = 315700
- 42 = 00101010
- 167 = 11000101
- => 1000101 0101010 = 8874

WAND – top-K query processing algorithm

$$S_{\text{TF-IDF}}(d, Q) = \sum_{t \in Q} tf_{d,t} \times \log \frac{N}{df_t}$$

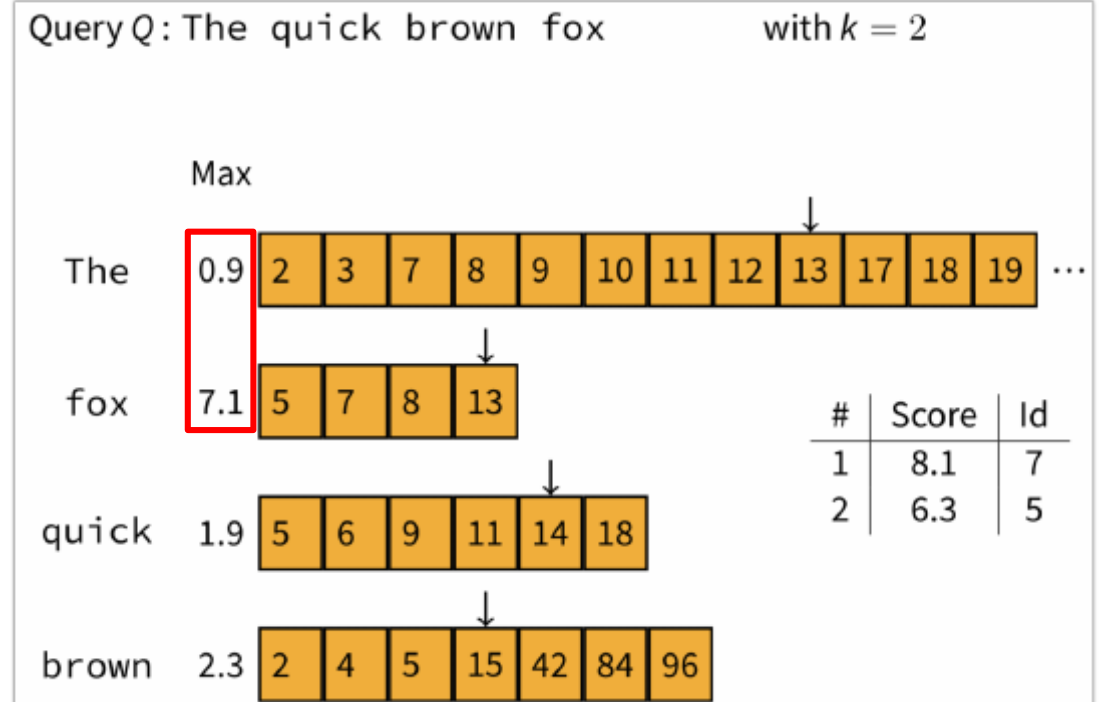
Query Q: The quick brown fox

with $k = 2$



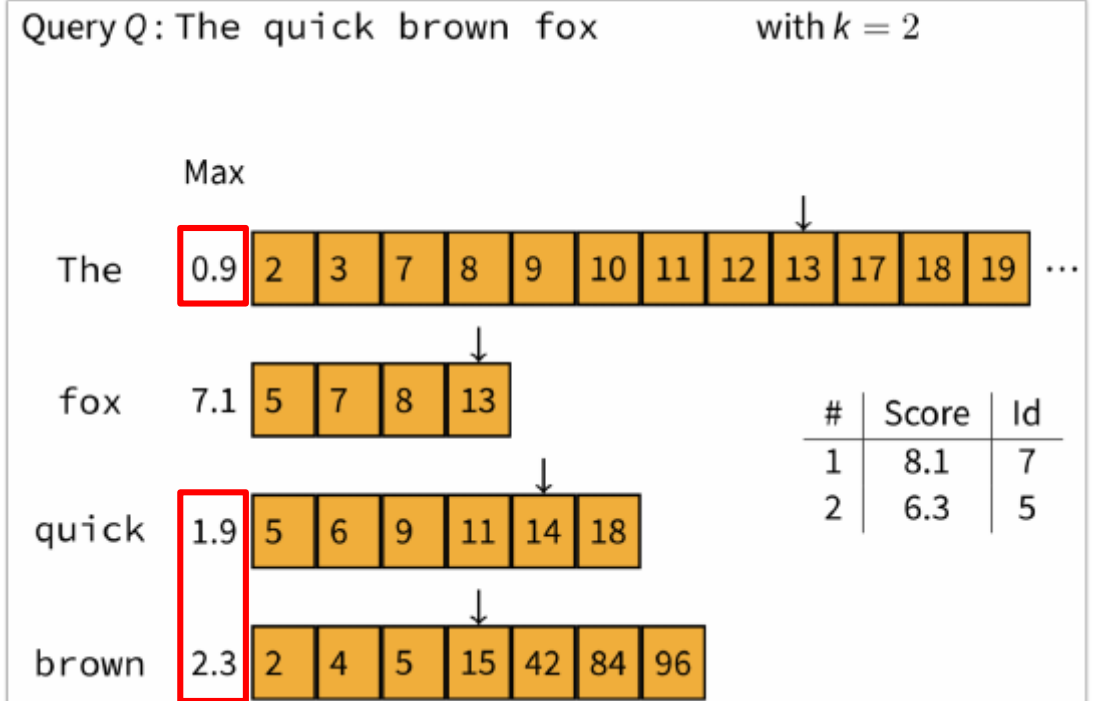
Q2

- Doc 13 is evaluated
- Max score for Doc 13 is $0.9 + 7.1 = 8.0$
- So Doc 13 might enter the top-2 list



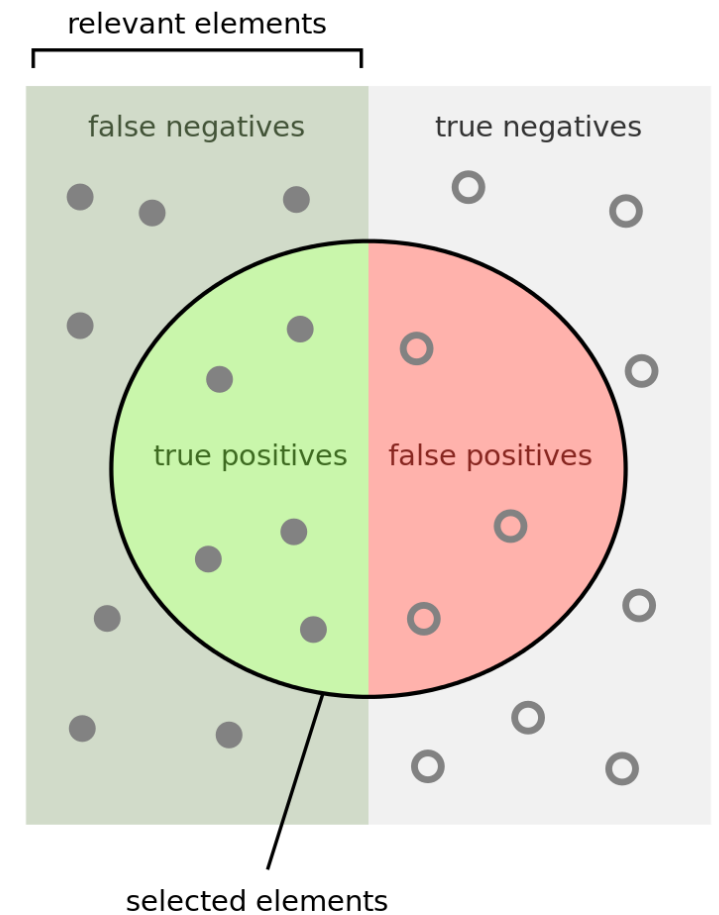
Q2

- No more documents will have “fox”
- Max score possible for a document with “The”, “quick” and “brown” will be $0.9 + 1.9 + 2.3 = 5.1$
- Lower than the scores of the top-2 documents
- So, we stop.



Recall and Precision

– https://en.wikipedia.org/wiki/Precision_and_recall



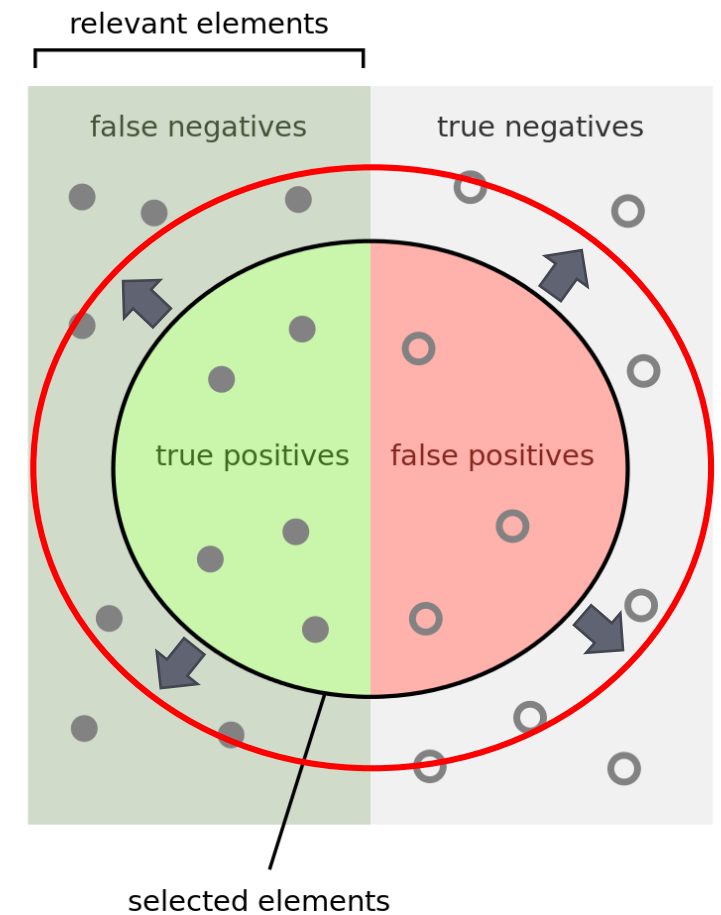
How many selected items are relevant?

$$\text{Precision} = \frac{\text{true positives}}{\text{true positives} + \text{false positives}}$$

How many relevant items are selected?

$$\text{Recall} = \frac{\text{true positives}}{\text{true positives} + \text{false negatives}}$$

Query Expansion



How many selected items are relevant?

$$\text{Precision} = \frac{\text{true positives}}{\text{true positives} + \text{false positives}}$$

How many relevant items are selected?

$$\text{Recall} = \frac{\text{true positives}}{\text{true positives} + \text{false negatives}}$$



Q4

- (a) User relevance feedback
 - e.g. ask users to click
- (b) Pseudo relevance feedback
 - e.g. blind feedback, search the top-K documents and perform topic modeling
- (c) Indirect relevance feedback
 - e.g. analyze query click logs to re-rank documents



Very Useful Online Resources

- Andrei Broder - WAND Revisited
- <https://youtu.be/gwsWUPVtt6Q?t=433>